Efficient Detection of Communication-related Performance Anti-patterns in Microservices

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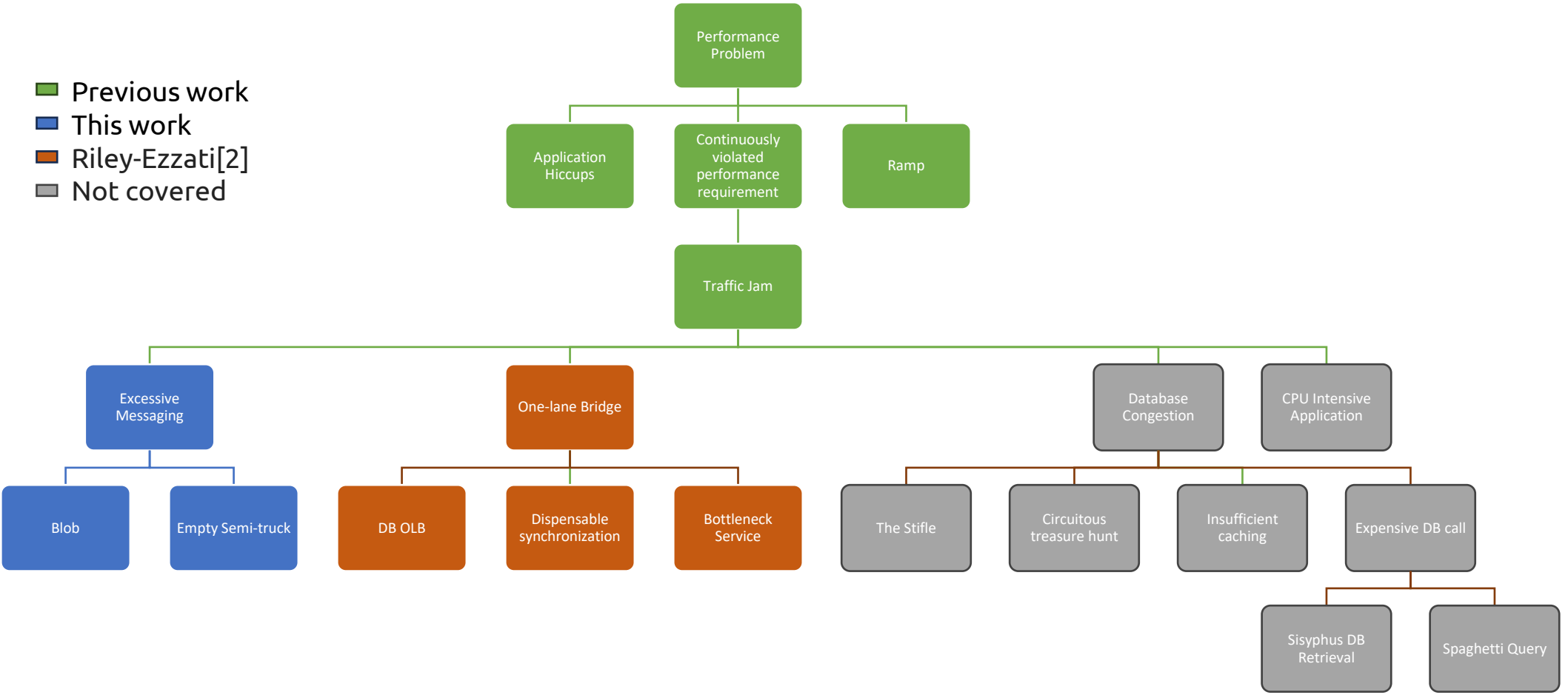
Introduction

- Performance anti-patterns (SPAs)
  - Bad practices which result in performance degradation
  - May not cause system failure, but impact performance



Agenda

- Problem statement
- Proposed approach and detection method
- Experiments
- Results
- Discussion and future work

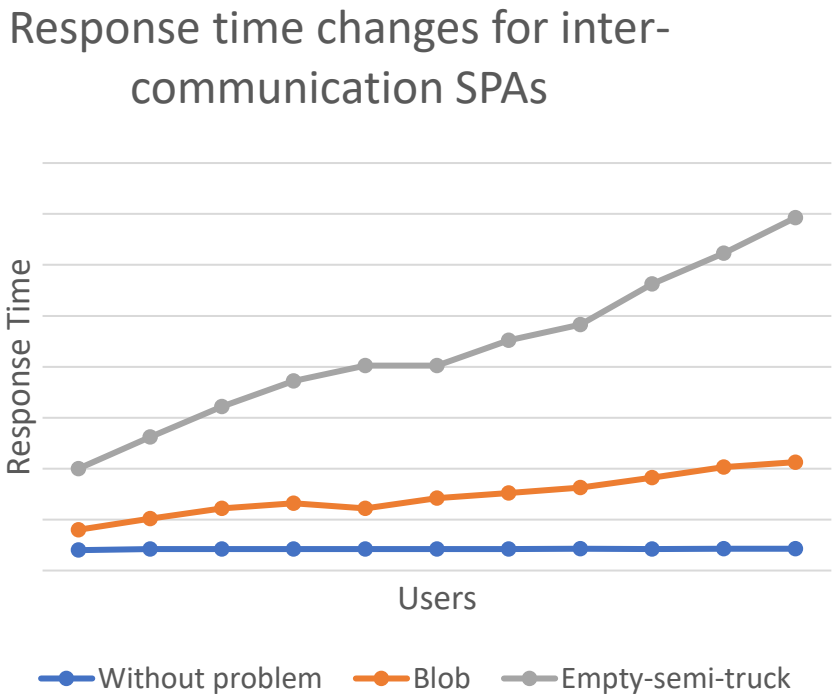
Problem Statement

Detecting Inter-Communication Performance Anti-Patterns in Microservice Environments

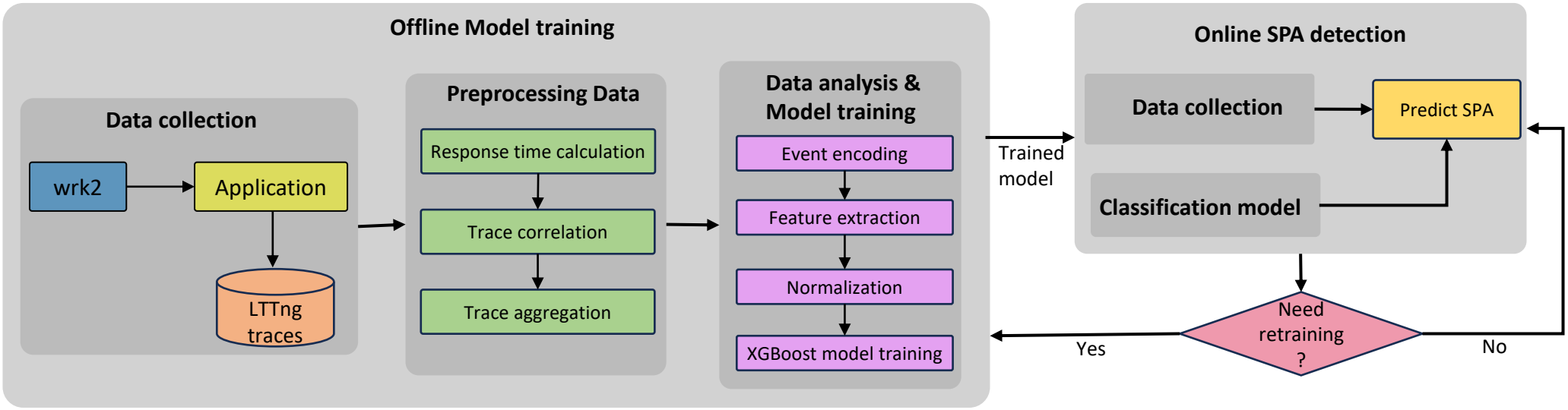
- **Non-intrusive method** for detecting anti-patterns in microservice-based applications.
- **Low overhead**: Minimal cost from trace data collection.
- **Improved accuracy** compared to existing detection methods.
- **Automated pipeline**, easily integrable with CI/CD processes.

Problem statement

- Blob
  - High communication overhead through remote communication between central blob component and other components
- Empty Semi-truck
  - Large number of small messages transmitted between two components as part of a single user request



Proposed Approach

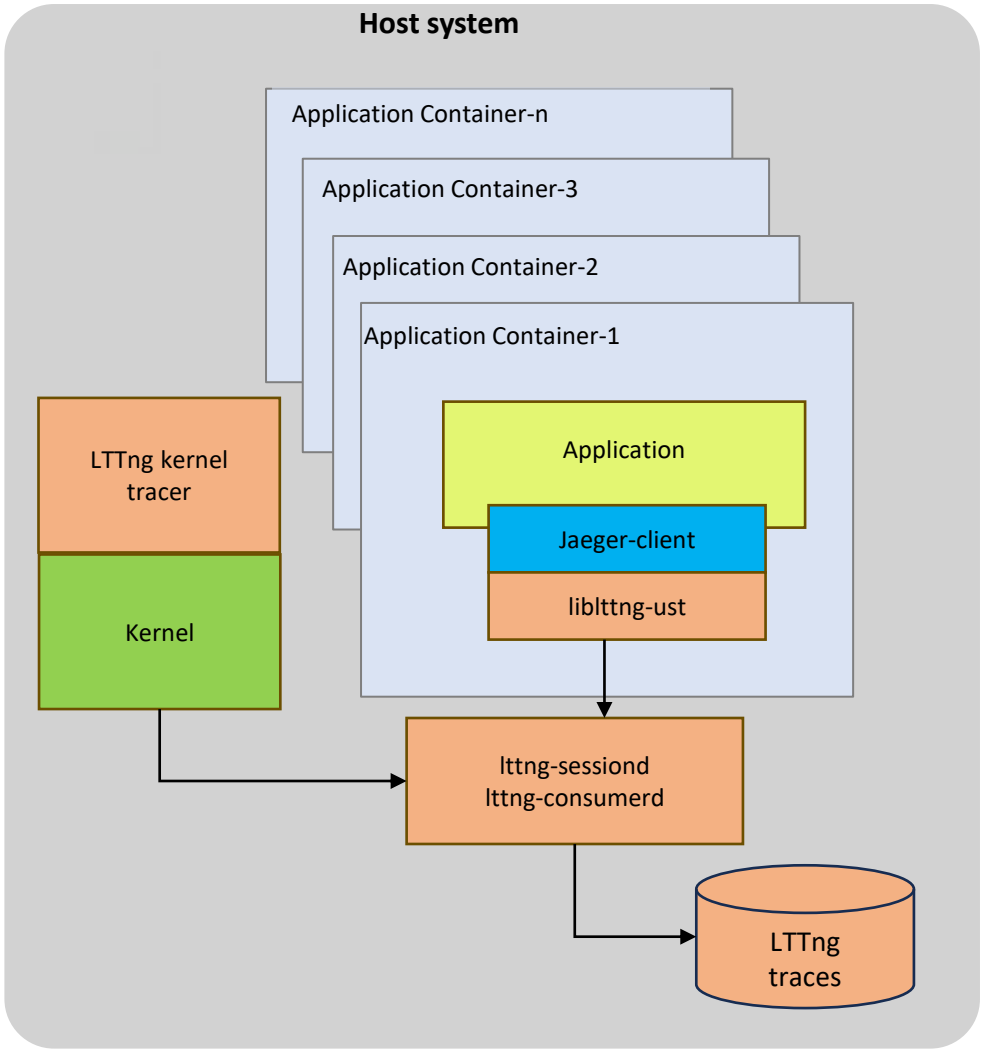


Proposed Approach- Hybrid Tracing

UST tracepoints on jaeger:

```
tracepoint(jaeger_ust, start_span, ctx.traceID().high(),
 ctx.traceID().low(), ctx.spanID(), ctx.parentID(), operationNameCStr,
serviceName().c_str(), startTimeSystemInt);

 tracepoint(jaeger_ust, end_span, _context.traceID().high(),
     _context.traceID().low(), _context.spanID(), duration);
```



Sample user-space and System call traces

Trace collection script:

lttng enable-event --userspace 'jaeger*' --channel=ust-channel --session test-session
lttng enable-event -k --channel=kernel-channel –syscall recvfrom, recvmsg, recvmmsg, sendto, sendmsg, sendmmsg

OpenTracing Trace sample:

Timestamp Channel CPU Event type Contents TID Prio PID Source Binary Location Function Location Source
01:24:47.480 957 960 ust-channel_6_0 6 jaeger_ust:start_span trace_id_high=0, trace_id_low=5485338799412484790, span_id=2723353438606276410, parent_span_id=2205893957549611025, op_name="UploadUserWithUserId", service_name="user-service", start_time=172646428748094910 6, context.packet_seq_num=0, context.cpu_id=6, context._vtid=111, context._vpid=1 111 1

System-call Trace sample:

Timestamp Channel CPU Event type Contents TID Prio PID Source Binary Location Function Location Source 21:44:27.237 576 630 kernel-channel_7_0 7 syscall_exit_recvfrom ret=12, ubuf=140192631653072, addr=0, addr_len=0, context.packet_seq_num=0, context.cpu_id=7, context._tid=3487992, context._vtid=52, context._pid=6076, context._vpid=1 3487992 6076 [net/socket.c:0]

Preprocessing- Aggregated trace

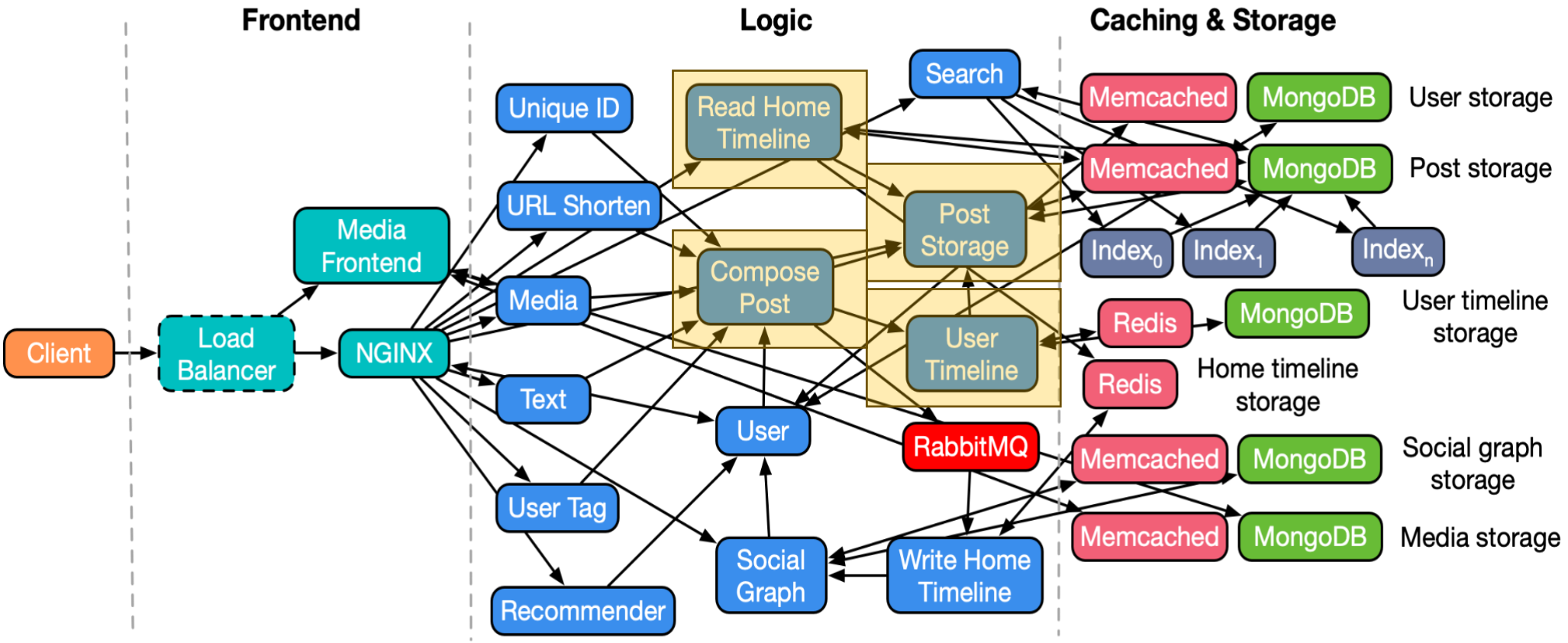
#. name,cur_ts,ret,response_time,span_id,parent_span_id,op_name,service_name,trace_id
1. jaeger_ust:start_span,1726464134589066276,0,1709015,4706249999899046019,17642707676433399526,ReadUserTimeline,user-timeline-service,9616759882794086189
2. jaeger_ust:start_span,1726464134589081297,0,24775,17102146431855677904,4706249999899046019,RedisFind,user-timeline-service,9616759882794086189
3. jaeger_ust:start_span,1726464134589469928,0,16601,16245312749159461 97,4706249999899046019,MongoFindUserTimeline,user-timeline-service,9616759882794086189
4. syscall_entry_sendmsg,1726464134589559979,195,54444,Null,Null,syscall_entry_sendmsg,user-timeline-service,9616759882794086189
5. syscall_entry_recvfrom,1726464134589633523,0,15908,Null,Null,syscall_entry_recvfrom,user-timeline-service,9616759882794086189
6. syscall_entry_recvfrom,1726464134590166434,116,19280,Null,Null,syscall_entry_recvfrom,user-timeline-service,9616759882794086189
7. syscall_entry_recvfrom,1726464134590390202,4,33007,Null,Null,syscall_entry_recvfrom,user-timeline-service,9616759882794086189
8. syscall_entry_recvfrom,1726464134590456198,123,27597,Null,Null,syscall_entry_recvfrom,user-timeline-service,9616759882794086189
9. jaeger_ust:start_span,1726464134590526296,0,19078,14929287647722053771,4706249999899046019,ReadPosts,post-storage-service,9616759882794086189

Data Analysis and Model Training

- Event encoding and padding
- Feature Extraction
  - Total Response Time, Maximum Response Time, Mean Response Time
  - Total Operations
  - Total Return Values, Maximum Return Value, Mean Return Value
  - Total Service Calls
  - Unique Operation Counts
- Feature Normalization
- Model Training with XGBoost
- Model validity checks
  - Supervised learning (classifier uses labeled trace data)
  - Feature information gain analysis
  - K-fold cross validation

Experiments

- TestBed:
  - DeathStarBench- Social Network application



SPA Injection

Performance anti-pattern	Impacted services	Details
Blob-1	UserTimeline, PostStorage	Moved whole logic of writeUserTimeline operation to PostStorage service
Blob-2	UserTimeline, PostStorage	Moved whole logic of readUserTimeline and writeUserTimeline operation to PostStorage service
Blob-3	UserTimeline, PostStorage, HomeTimeline	Moved whole logic of readUserTimeline, writeUserTimeline, and readHomeTimeline operations to PostStorage service
Empty-semi-trucks-1	ComposePost	In ComposeAndUpload commit each call separately to simulate overhead
Empty-semi-trucks-2	UserTimeline	In readUserTimeline instead of fetching all posts at once, the function uses a for loop where each post is queried from MongoDB individually

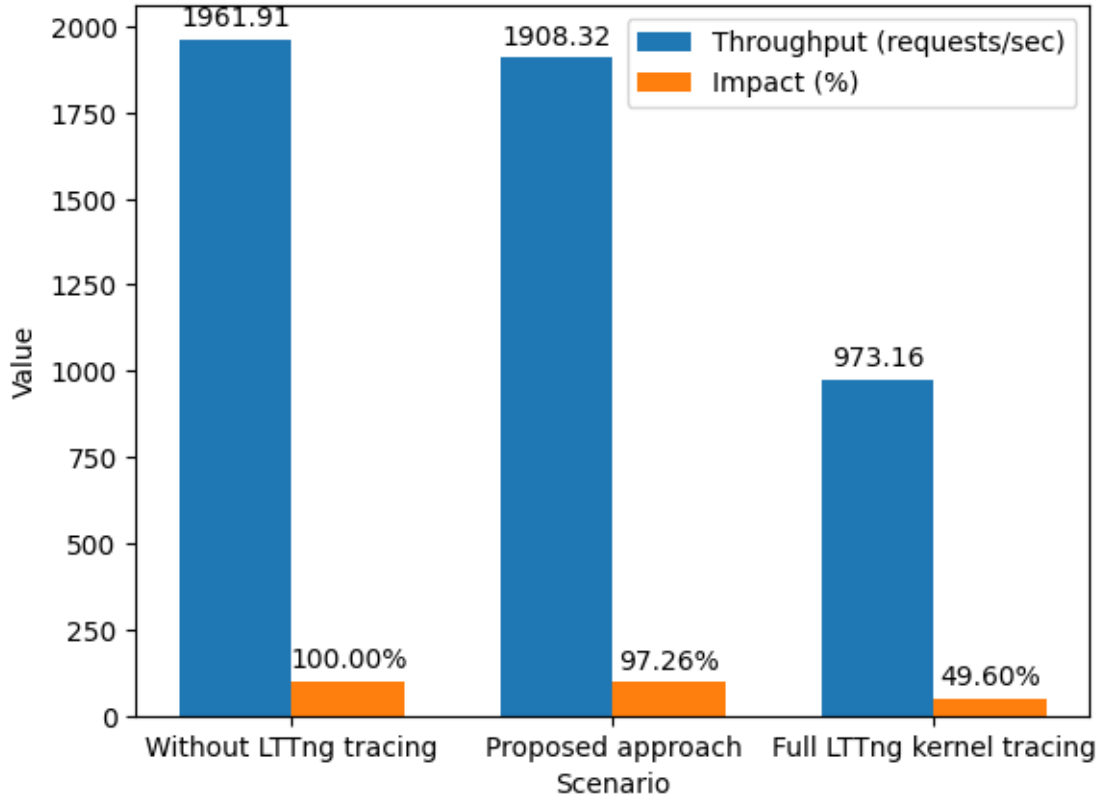
Experiments and Results

#	Scenario	Details
1	Baseline	unchanged application
2	Baseline with system faults	unchanged application with noise
3	Blob-1	Blob-1 injected
4	Blob-2	Blob-2 injected
5	Blob-3	Blob-3 injected
6	Empty-semi-trucks-1	Empty-semi-trucks-1 injected
7	Empty-semi-trucks-2	Empty-semi-trucks-2 injected

Comparison of using different classifiers to detect Blob and Empty-semi-trucks SPAs

Classifier	Precision	Recall	Accuracy
SVM	72.80	72.87	99.81
XGBoost	73.31	73.26	99.98
Random forest	73.31	73.25	99.98

Overhead analysis



Contributions

1. Detecting SPAs by using a non-intrusive, low-cost tracing method.
2. Detection based on selective syscall traces, specifically related to the intended SPA.
3. Benefiting from existing open-tracing instrumentation in the microservice applications.
4. Low-cost SPA detection approach
5. Use of distributed traces jointly with system calls to provide a better characterization of the normal behavior of the system vs anti-pattern effect.
6. Facilitating root-cause analysis
7. Presenting a modern classification method for SPAs, with high detection accuracy (99%)
8. Considering calls between micro-services in the detection model training
9. Advanced baseline calculation method by calculating baseline for each trace type

Future directions

- Using methods like topic modeling to decrease preprocessing efforts
- Add more experiments to build a dataset of performance anti-patterns

References

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Thanks for your attention

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